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3 **Warming of the Indian Ocean threatens eastern and southern Africa, but could
4 be mitigated by agricultural development**

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1 **Abstract**

2
3 Since 1980, the number of undernourished people in eastern and southern Africa has more than doubled. Rural
4 development stalled and rural poverty expanded during the 1990s. Population growth remains very high and
5 declining per capita agricultural capacity retards progress towards Millennium Development goals. Analyses of
6 in situ station data and satellite observations of precipitation identify another problematic trend. Main growing
7 season rainfall receipts have diminished by ~15% in food insecure countries clustered along the western rim of
8 the Indian Ocean. Occurring during the main growing seasons in poor countries dependent on rain fed
9 agriculture, these declines are societally dangerous. Will they persist or intensify? Tracing moisture deficits
10 upstream to an anthropogenically warming Indian Ocean leads us to conclude that further rainfall declines are
11 likely. We present analyses suggesting that warming in the central Indian Ocean disrupts onshore moisture
12 transports, reducing continental rainfall. Thus late 20th century anthropogenic Indian Ocean warming has
13 probably already produced societally dangerous climate change by creating drought and social disruption in
14 some of the world's most fragile food economies. We quantify the potential impacts of the observed
15 precipitation and agricultural capacity trends by modeling millions undernourished people as a function of
16 rainfall, population, cultivated area, seed and fertilizer use. Persistence of current tendencies may result in a
17 50% increase in undernourished people. On the other hand, modest increases in per capita agricultural
18 productivity could more than offset the observed precipitation declines. Investing in agricultural development
19 can help mitigate climate change while decreasing rural poverty and vulnerability.

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21

22

1 **Text**
2

3 **Introduction – food insecurity along Africa’s eastern coast**
4

5 Sustainability science often looks at trends and the interaction of trends(1, 2) through the lens of
6 ‘use-inspired basic research’(1). We present here an analysis of rainfall and agricultural capacity
7 trends inspired by our research supporting earlier early warning for the US Agency for
8 International Development’s Famine Early Warning System (Table S1). Most of this work has
9 focused on monitoring and predicting emerging drought conditions in Eastern and Southern
10 Africa(3-5), one way of improving adaptive capacity(6) recommended by the Intergovernmental
11 Panel on Climate Change (IPCC). In 2003, however, a routine attempt to use early season rainfall
12 (March-April) rainfall as a predictor of late season (September-October) crop production produced
13 a disturbing discovery. Substantial rainfall declines had occurred within critical crop growing areas
14 in Ethiopia. The place and timing of these trends made them dangerous. Further analysis suggested
15 links to drought across much of eastern eastern Africa during the long rains season, and potential
16 anthropogenic links to the central Indian Ocean(3, 5). In addition to climate analysis, this work
17 also included evaluations of population, agriculture and water security. We find that in eastern and
18 southern Africa, the combination of declining per capita agricultural capacity(7) and increasing
19 aridity is exacerbating vulnerability(8) and rural poverty(9). Declining investments in rural
20 development, rapidly increasing rural populations, the depletion of soil nutrients through erosion,
21 and the cultivation of most cultivatable areas limit agricultural productivity growth(10). Low per
22 capita agricultural production and rural poverty go hand in hand, slowing progress towards
23 Millennium Development Goals (MDGs). A recent overview of these goals(11) indicates
24 differential progress on different fronts. Education and economic conditions have improved.
25 Population growth remains high (2.3 percent a year). Progress on hunger and health has been
26 halting. Our concern is that post-1980 rainfall declines during the main eastern African long rains
27 (March to May) and the main southern African summer rains (December to February) may be
28 contributing to African food insecurity. We believe that these declines may be associated with
29 anthropogenic warming in the Indian Ocean. If so, these precipitation declines, acting in concert
30 with declining agricultural capacity, likely constitute a prime example of societally dangerous
31 climate change.

32
33 In this report we quantify the likely impact of observed precipitation and agricultural capacity
34 trends using empirical food balance indicator models. These models estimate the millions of
35 undernourished people as a function of population, cropped area, seed use, fertilizer applications,
36 and main growing season precipitation. These models show that continued declines in rainfall and
37 per capita agricultural capacity will produce increasing food insecurity. We next use observations
38 and climate model simulations to argue that recent declines in eastern and southern African
39 growing season rainfall are linked to anthropogenic warming in the Indian Ocean. This link to
40 global warming implies that these precipitation declines are likely to continue or intensify. For
41 eastern Africa, this result is at odds with the most recent IPCC assessment(12, 13), which
42 anticipates precipitation increases. For southern Africa, this result is consistent with previous
43 analyses(13, 14), which anticipate rainfall declines. We are essentially arguing that warming in the
44 south-central Indian Ocean (0-15°S, 60-90°E) has a similar effect on near coastal eastern and
45 southern Africa during boreal winter and spring. In each season, oceanic warming appears to
46 reduce onshore moisture transports(5, 15-17) while increasing continental atmospheric stability.

1 This may explain recent drought tendencies in gauge data, satellite-observed precipitation*, lake
2 levels(5, 16) and vegetation indices in densely populated water-insecure regions of eastern
3 African. If a common anthropogenic Walker-cell like disruption has caused declines in both
4 southern and eastern African rainfall, then this disruption is likely to persist. Furthermore, recent
5 climate change impact assessments(18), based on optimistic precipitation simulations over eastern
6 Africa, may underestimate yield reductions(4). Our food balance modeling of millions of
7 undernourished, however, suggests that these impacts could be mitigated by agricultural
8 development.

9

10 Modeling the impacts of agricultural capacity and rainfall

11 Over the past 25 years the population of food insecure eastern and southern Africa has doubled
12 while per capita cropped area has declined by 33% and the millions of undernourished increased
13 by 80%(19). Today, one out of three small children is dangerously underweight and 40% of 308
14 million people are undernourished. Average national per capita cropped area values are often
15 below 1,000 m² per person[†]. Food aid has become chronic, with 2005 World Food Programme
16 (WFP) aid valued at \$200US million. Poor farmers in these countries, often trapped in cycles of
17 displacement, division and degradation(20), depend on rainfed agriculture(21). Per capita crop
18 production is an important metric of food availability and security. Limited technological inputs
19 create a strong dependence between national average cropped area and national average production
20 ($r=0.95$).

22 Recent agricultural capacity trends tend to be dominated by cropped area and population growth
23 increases*. In general, increased food production has matched growing population through
24 increased labor inputs and through extensification, with area under cultivation increasing by 50%
25 over the past 25 years while population has almost doubled (Fig. 1.A). Seed and fertilizer use has
26 also lagged population growth. Unfortunately, since 1990, overseas agricultural assistance has
27 declined from 12 to 4% of total foreign aid(7), and only 4% of African public spending goes to
28 agriculture(7). This has contributed to stalling increases in agricultural inputs. As the gap between
29 population growth and structural agricultural components continues to grow, vulnerability and
30 rural poverty will increase, amplifying the impact of agricultural droughts. These droughts appear
31 to be more severe in recent years (Fig. 1.A), and main growing season rainfall* has declined by
32 about 15% across Eastern and Southern Africa, in step with recent increases in radiative
33 forcing(22). The interaction of agricultural and rainfall trends may be represented by empirical
34 models estimating the millions of undernourished people.

36 The five components of national per capita production (rainfall, population, cropped area, seed and
37 fertilizer use) can be combined into two variables: rainfall and per capita agricultural capacity.
38 Agricultural capacity represents the slowly varying non-weather component of a national food
39 balance, and is a function of cropped area, seeds and fertilizer divided by population with units of
40 per capita crop production [kg person⁻¹ year⁻¹]. Fig. 1.B shows main growing season rainfall and
41 per capita agricultural capacity trends for ten eastern and southern African nations. Somalia has
42 been excluded due to limited data. Downward point arrows (Ethiopia, Kenya, Burundi, Tanzania,

* Please see our online supporting materials for a specific description of our methods.

† Agriculture, population, food security and food aid statistics were obtained from the UN Food and Agriculture Organization (FAO)

1 Malawi, Zambia, Zimbabwe) indicate countries experiencing key growing season rainfall declines
2 between 1979 and 2005. Note the geographic clustering of observed rainfall declines along
3 Africa's eastern seaboard. Countries with left pointing vectors (Ethiopia, Uganda, Zambia,
4 Zimbabwe) have experienced 1994-2003 declines.

5
6 In these semi-arid countries a strong dependence on rain-fed smallholder farming practices results
7 in quasi-linear relationships between grain yields, seasonal rainfall receipts, and food deficits.
8 Hence, agricultural capacity multiplied by percent normal rainfall is strongly related to per capita
9 production. The inverse of this measure (food imbalance) is related to food aid*. For each country
10 (except Ethiopia, due to data limitations), the food imbalance measure was regressed against
11 'millions of undernourished' statistics obtained from the United Nations' Food and Agriculture
12 Organization. This gives us a pragmatic means of translating changes in rainfall, cropped area,
13 seed and fertilizer use into an index of food insecurity, grounded empirically by historic
14 observations. The resulting model performed well at a national/inter-annual scale for eastern and
15 southern Africa ($r=0.86$, $p=1.6 \times 10^{-5}$). In order to stress the semi-quantitative nature of these
16 projections, we present our results as percent changes from a 2000 baseline. This also allows
17 comparison with the Millennium Development goal of halving the number of undernourished by
18 2015.

19
20 Three scenarios were examined (Fig. 1.C). Scenario 1 represents a 'business as usual' future.
21 Recent precipitation and agricultural capacity trends were assumed to persist through 2030. Under
22 this scenario, the aggregate tendencies shown in Fig. 1.A continue, and undernourishment
23 increases by 53% between 2000 and 2030. The effects of agricultural capacity declines alone can
24 be seen in Scenario 2. Undernourishment increases by 23%. On the other hand, evaluating a
25 combination of observed rainfall declines paired with moderate increases in per capita agricultural
26 capacity of two kilograms per person per year, equivalent to a 2% per capita growth rate, exhibited
27 substantial (38%) declines in undernourishment. Drought interacts dangerously with low
28 agricultural capacity, but its effects can be mitigated through agricultural development.

29

30 **Walker cell-like anomalies may explain observed rainfall declines**

31

32 Will the observed rainfall declines (Fig. 1) persist, increasing levels of undernourishment (Fig.
33 1.C)? Fueling this concern is the strong covariance between global temperatures ($r=0.82$) and
34 south-central Indian Ocean ($0-15^{\circ}\text{S}$, $60-90^{\circ}\text{E}$) sea surface temperatures (SSTs). Both have risen
35 dramatically since the 1980s, along with radiative forcing (Fig. 2.A), and tropical Indian Ocean
36 SSTs recently reached their highest value in 120,000 years(23). While anthropogenic warming has
37 occurred in all oceans(24), this warming has been larger in the Indian Ocean(25). Recent careful
38 analysis identifies an $0.5-1.0^{\circ}\text{C}$ 1960-1999 SST increase in the Indian Ocean Thermal
39 Archive(26). This study also finds that most (7 out of 10) World Climate Research Programme
40 Coupled Model Intercomparison Project(27) (CMIP3) models recreate this tropical warming
41 tendency, strongly implicating greenhouse gas and aerosols emissions in the recent observed
42 warming (Fig. 2.D). Modelled CMIP3 increases in southern tropical Indian Ocean temperatures
43 are strongly linked to trends in equatorial Pacific zonal wind stress and southward shifts in the
44 latitudinal position of the subtropical gyre. Warmer oceanic SSTs have produced substantial
45 increases in precipitation over the tropical Indian Ocean, indicated by both satellite(28) (13%) and
46 reanalysis(29) (28%) datasets. This convection releases energy in the atmosphere, influencing the

1 regional weather via a Walker cell-like circulation anomaly. This mechanism may have been
2 responsible for the 1984 Ethiopian drought(30).

3
4 The statistical connection between increased oceanic precipitation and continental rainfall is
5 shown in Fig. 2.B^{*}. Both data sources have been normalized, hence a regression coefficient of -0.6
6 associates a one standard deviation increase in oceanic precipitation with a -0.6 standard deviation
7 decrease in African rainfall. The latitude of negative associations shifts with the march of seasons,
8 following the thermal equator and path of maximum onshore moisture transports (shown
9 schematically with arrows).

10
11 The connection between seasonal moisture transports and central Indian Ocean warming
12 (represented by precipitation) can be explored empirically via canonical correlation analysis
13 (CCA). CCA is a statistical technique that finds dominant patterns of covariability in two sets of
14 multivariate data. The two sets of data examined were seasonal 1950-2005 zonal reanalysis
15 moisture transports over Africa's eastern seaboard (20°N - 20°S , 25°E - 45°E) and seasonal tropical
16 Indian Ocean precipitation (15°N - 15°S , 55°E - 90°E). Tropical oceanic precipitation is linearly
17 related to diabatic atmospheric heating, which can produce east-west Walker cell-like circulations
18 anomalies(31). Our hypothesis is that these anomalous circulations control decadal rainfall
19 fluctuations endangering food insecure Africa.

20
21 Fig. 2.C shows our CCA results for March-April-May (other seasons are presented in Fig. S1).
22 In each season, we find that the 1st canonical correlate identifies a 1st CCA pattern describing
23 shared Indian Ocean precipitation and moisture transport variations. This approach produces time-
24 series describing covariant Indian Ocean precipitation (CC_P) and moisture transports (CC_M).
25 Increasing tropical Indian Ocean precipitation (blue circles) is associated with atmospheric ridging
26 and anti-cyclonic moisture circulations over eastern Africa (arrows). This disrupts the main
27 onshore moisture flows (shown with blue shading). University of Delaware precipitation deficits
28 for the strongest six oceanic precipitation events are shown with red boxes. While only boreal
29 spring results are shown here, the patterns for other seasons are similar; a Walker cell-like dipole
30 emerges as the dominant (1st canonical correlate) relationship. The transport correlations indicate
31 reduced onshore moisture fluxes between 0 and 15°S . This region supplies 66% of Indian Ocean-
32 sourced zonal moisture transports into Tropical Africa, and 83% of the annual mean zonal
33 moisture transports entering tropical Africa originate in the Indian Ocean. The northern flanks of
34 the CCA patterns are associated with transport anomalies flowing from the drier Sahelian regions.
35 Thus changes in central Indian Ocean precipitation are correlated with transitions between dry
36 continental and moist oceanic air masses over tropical Africa.

37
38 At low (decadal) frequencies, correlations between the CC_M and CC_P approach unity, and the
39 zonal transport indices are negatively correlated ($r \sim -0.9$) with regional rainfall (Table 1). In total
40 the CCA-identified transports can explain 85% of the low frequency variations in eastern MAM,
41 southern DJF and Sahelian JJA rainfall. All three seasons/regions experienced a 1960-1970 decline
42 in observed rainfall and CCA transport forcing, in sync with anthropogenic 1960-1970 increases
43 in Indian Ocean heat content(24-26). A second heat content increase in the early 1980s was
44 associated with MAM and DJF transport disruptions and drying tendencies in southern and eastern
45 Africa. Slower increases in JJA transport disruptions, combined with north/south shifts in Atlantic

1 sea surface temperatures (SSTs), has likely led to a modest recovery in Sahelian rainfall(14, 32),
2 albeit with continued declines across the eastern Sahel and southern Ethiopia(32).
3

4 To examine the influence of diabatic heating on moisture transports in a simulation framework, we
5 have also conducted experiments with the NCAR CAM*, a fully non-linear atmospheric model
6 with explicitly resolved moist processes. An additional diabatic heating term is added over the
7 central Indian Ocean, which increases the local rainfall, as in Barlow et al(33). The added heating
8 is sinusoidal in latitude and longitude for a half period over the domain -20° S to 5°N and 50 to
9 100°E with a magnitude of 0.02 W kg⁻¹. This southern central Indian Ocean domain was chosen to
10 match our previous observational analyses(5, 16) and the early assertion that enhanced cyclonic
11 activity helped create the tragic 1984 Ethiopian Famine(30). A clear influence on moisture
12 transport over the interior of Africa is observed (Fig. 2.D), in good agreement with the
13 expectations of simple Gill-type dynamics, and is associated with decreases in local rainfall.
14 These MAM circulation anomalies are similar to previous modeling results for JFM(17), and the
15 observed low-level circulation anomalies associated with warm Indian Ocean warming seasons for
16 boreal spring (Fig. 2.C). Thus while the relevance of the simple Gill model might be questioned,
17 especially as it neglects the mean wind(34), the general link between increased central Indian
18 Ocean precipitation and decreased onshore moisture transport appears robust even in a full
19 atmospheric model.

20

21 In summary, the strong, probably anthropogenic(24-26) warming of the south-central Indian
22 Ocean (Fig. 2.A), increases maritime precipitation, which is statistically and dynamically related to
23 continental rainfall declines and moisture transport reductions (Figs. 2B, 2C). Non-linear climate
24 simulations driven by diabatic heating over the Indian Ocean produce similar transport anomalies
25 (Fig. 2.D and ref. (17)). In this context, what do projected 21st climate changes suggest over the
26 Indian Ocean and tropical Africa?

27

28 **Climate change simulations suggest continued Indian Ocean Warming**

29

30 In this section we present a multi-model analysis of CMIP3 precipitation simulations. If the
31 consensus of these models suggests continued increases in oceanic precipitation, then we should be
32 concerned that continued continental rainfall declines, similar to those shown in Fig. 2B, will lead
33 to increased undernourishment unless mitigated (Fig. 1.C). We selected 11 models having both
34 20th century (20c3m) and 21st century climate simulations. Because near-term (2000-2030)
35 projections of radiative forcing(22) are quite similar across the various emission projections(35),
36 the various 21st century scenarios were combined. Figure 3 summarizes results for the critical
37 March-April-May east African growing season*. December-January-February results were similar
38 and are not shown.

39

40 We begin by analyzing east-west (zonal) tropical precipitation tendencies. Panel 3.A shows a time-
41 longitude plot of temporal z-scores (standard deviations, σ) of the CMIP3 ensemble means
42 averaged between 15°N/S. In this plot the 1950-2100 MAM multi-model means were averaged
43 between 15°N and 15°S at 5° intervals between 20°E and 160°E. The time-series at each longitude
44 was translated into z-scores (standard deviations) and plotted with red (blue) indicating reduced
45 (enhanced) precipitation. As greenhouse gas forcing increases with time, beginning in the 1970s,
46 precipitation increases across the Indian Ocean and declines across tropical Africa. This timing is

1 broadly consistent with observed increases in oceanic SSTs (Figure 2.A) and reductions in
2 continental rainfall (Fig. 1.A). We can analyze scatter surrounding the CMIP3 climate change
3 consensus by translating the 132 individual precipitation simulations into time-varying probability
4 distribution functions *(36) (pdfs). Fig. 3.B shows the multi-model ensemble precipitation pdfs
5 over the tropical Indian Ocean (0-15°S, 60-90°E). As greenhouse gasses and aerosols accumulate
6 in the atmosphere, the Indian Ocean warms, oceanic precipitation increases, presumably producing
7 moisture transport variations and tropical African rainfall declines similar to those shown in Fig. 2.
8 The pdfs are tight and trend strongly, increasing steeply after the mid 1970s, in step with radiative
9 forcing increases(22) and observed African rainfall declines (Fig. 1) so that by late in the 21st
10 Century the ranges of precipitation projected are quite different from historical ranges. Panel 3.C
11 shows Indian Ocean precipitation changes for the eleven different individual models (x axis) along
12 with the correlation between each model's multi-simulation mean time-series and the ensemble
13 mean (y). All models exhibit positive correlations, so that the models were unanimous in
14 projecting increases in March-April-May atmospheric warming (i.e. precipitation) over the tropical
15 Indian Ocean. The frequency with which MAM central Indian Ocean precipitation totals in 2031-
16 2050 exceeded the average rainfall total for 1951-1970 increased by more than 20% in 10 of the 11
17 models. Taken together the panels of Fig 3 indicate tilt of the odds towards more precipitation over
18 the Indian Ocean and less precipitation over Africa's eastern seaboard.

19

20 **Summary and Conclusions**

21

22 Two independent, but interacting, post-1980 tendencies have contributed to food insecurity in
23 eastern and southern Africa (Fig. 1). First, population growth has exceeded increases in
24 agricultural infrastructure and cultivated area. Second, there has been a tendency for main growing
25 season rainfall to decline. Empirical and model-based explorations (Fig. 2, S2) support assertions
26 that Walker-cell-like disruptions to of atmospheric circulations and moisture transports link a
27 warming Indian Ocean to a drier eastern African seaboard. Multi-model CMIP3 ensembles (Fig. 3
28 and ref (17)) suggest greenhouse gas and aerosols emissions have contributed substantially to this
29 observed late twentieth century warming (Fig. 2.A). Anthropogenic climate change has probably
30 produced societally dangerous increases in eastern and southern African food security. These
31 conclusions, in general, agree with the most recent 4th Intergovernmental on Panel on Climate
32 Change (IPCC) finding that semi-arid Africa may experience large scale water-stress(37) and yield
33 reductions(18) by 2030.

34

35 The results presented here, however, differ in approach from the IPCC models, focusing on
36 empirical relationships rather than raw CMIP3 precipitation simulations. These raw simulations
37 have implied that eastern Africa might become wetter while southern Africa becomes drier³⁰. We
38 question the fidelity, in general, of continental precipitation simulations, especially over the
39 extremely complex terrain of east Africa. Global models suppress important local mechanisms
40 (internal thermal and orographic gravity waves³¹), and vary substantially in their ability to
41 represent the transient systems that draw kinetic energy from the mean circulation, overcoming
42 stability inhibitions and producing organized mesoscale convection³². Even monthly reanalysis
43 precipitation fields have almost no skill ($R^2 < 0.2$) when compared with observations in eastern
44 Africa. Given this uncertainty, we believe that the observed quasi-linear negative relationships
45 between central Indian Ocean warming and east African rainfall represents the mostly likely
46 outcome. This may be indicative of other 'Indian Rim' and South American countries as well,

1 since similar precipitation reformulations also suggest main season declines(38), with the result
2 that main growing season droughts may disproportionately affect tropical and sub-tropical
3 countries. Global assessments of anthropogenic precipitation(13) and yield(18) changes may be
4 underestimating these drought signals. The climate change impacts in low-income nations have
5 been almost completely driven by emissions from middle and high income countries(39). While
6 more research, and better climate models, will be required to settle quantitatively issues of
7 attribution, the social and economic disruptions associated with anthropogenic drought may add
8 substantially to these impacts.

9
10 Vulnerability arises through a complex interplay of exposures, sensitivities, and resiliencies that
11 can either dampen or amplify the impact of climatic shocks(8). The recent IPCC assessment
12 identifies many means of enhancing adaptive capacity(6) in Africa. Improved forecasts and earlier
13 early warning can help(4), as can improving existing rain-fed agriculture through farm
14 management(40) (Cf. Table 2 in Ref (6)), as can enhanced bio-technological(41) inputs. Recent
15 research suggests that investments in African agricultural development will lead to substantial pro-
16 poor economic growth(41-43). The continuation of recent tendencies will be likely to result in
17 continued increases in undernourishment. This is not a *fait accompli*. The range of possible human
18 outcomes is large, and depends largely on the choices that we make. By 2025 rates of child
19 malnutrition, for example, could range from 9.4 to 41.9 million(44). Our food aid models suggest
20 that a 15% increase in yields per decade (equivalent to a 2 kg per person per year increase in
21 agricultural capacity) could come close to achieving the MDG of halving the number of
22 undernourished, albeit by 2030 rather than 2015.

23
24 While notable achievements in economic growth and education have been made since 1999, recent
25 progress on hunger in Africa has been very slow(11) and has fallen short of keeping pace with the
26 relatively rapid population growth. The past few years have seen an increased governmental
27 appreciation of the role of agricultural development(7), but public investments in agriculture and
28 international donor assistance have lagged behind MDG commitments(11). More and higher
29 quality assistance and agricultural policies, guided by objective policy design and better
30 governance, can help eastern and southern Africa achieve sustainability transition. Investments in
31 agricultural capacity seem warranted, given the large financial and humanitarian costs of business
32 as usual in a changing climate system. These investments may also be just, given the strong
33 evidence for anthropogenic drought and the disproportionate emissions by middle and high income
34 countries(39).

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5 Model Diagnosis and Intercomparison (PCMDI) for collecting and archiving the model output, and the JSC/CLIVAR
6 Working Group on Coupled Modeling (WGCM) for organizing the model data analysis activity. The multi-model
7 data archive is supported by the Office of Science, U.S. Department of Energy.

8 **Figure legends**

9 **Fig. 1. Food security analysis.** **A.** Rainfall, population, cropped area, seed use and fertilizer use
10 for eastern and southern Africa. Rainfall is expressed as a percent of the 1951-1980 average. The
11 other variables are expressed as percents of 1979-1981 averages. Global radiative forcing is shown
12 with a stippled line on an inverted axis. **B.** The black vectors denote recent observed rainfall and
13 agricultural capacity tendencies. Agricultural data were obtained from the Food and Agriculture
14 Organization. **C.** Empirical food balance models results, expressed as percentages of 2000
15 undernourishment.

16 **Size: 2 column (18 cm x 4.3 cm)**

17 **Fig. 2. A.** time-series of GISS global temperatures and Kaplan Indian Ocean SSTs. **B.** Pixel-by-
18 pixel bivariate regressions coefficients for seasonal standardized precipitation indices. For each
19 season standardized southwest Indian Ocean (SWIO) GPCP precipitation. The red arrows show
20 the general location of lower tropospheric (850 hPa) westward wind maxima. **C.** First canonical
21 correlates for each season. Results have been screened for significance ($p < 0.1$). The blue shading
22 denotes the core of the westward transports, with mean 850 hPa wind speeds of greater than 4 ms^{-1} .
23 **D.** CAM diabatic forcing expirement results. Arrows indicate simulated moisture transport
24 anomalies [$\text{kg kg}^{-1} \text{ms}^{-1}$]. Blue-red shading indicates diabatic heating anomalies [K day^{-1}].

25 **Size: 2 columns (18 cm x 11 cm)**

26 **Fig 3. CMIP3 March-April-May precipitation simulation results.** **A.** A time (1950-2100)
27 versus longitude plot of CMIP3 ensemble means averaged between 15°N/S . At each 5° longitude
28 box the 150 years of precipitation values were translated into z-scores (standard deviations, σ).
29 Red shades indicate drier conditions. Blue shades indicate above normal precipitation. **B.** A time
30 versus probability plot showing the multi-model ensemble pdfs²⁹ for the south-central Indian
31 Ocean ($0-15^\circ\text{S}$, $60-90^\circ\text{E}$). **C.** Summary of changes in frequency of above “normal” precipitation
32 and correlation of precipitation changes in individual climate models with the multi-model
33 ensemble means for this same region.

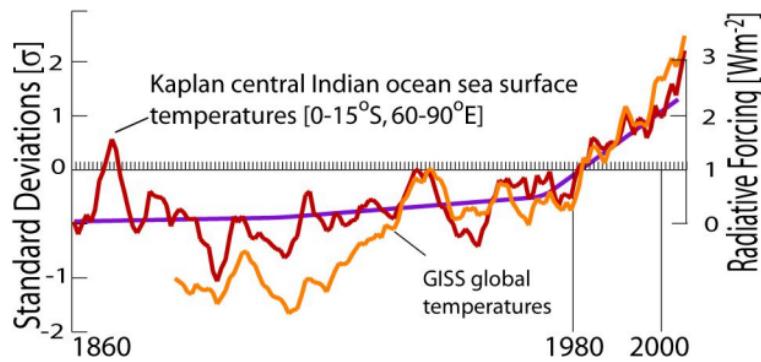
34 **Size: 2 columns (18 cm x 6 cm)**

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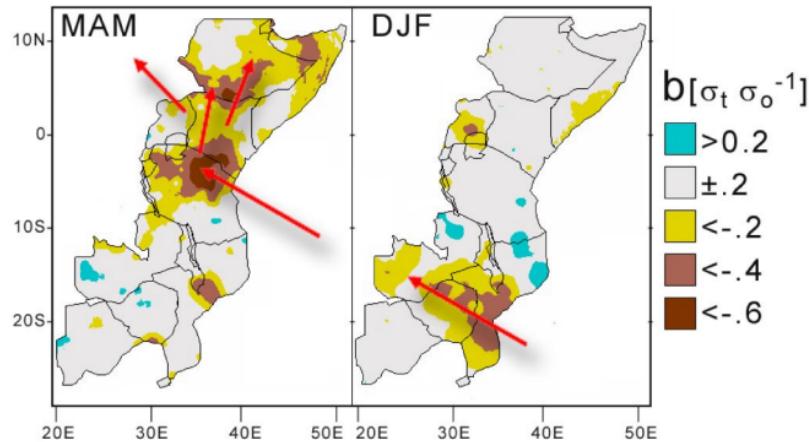
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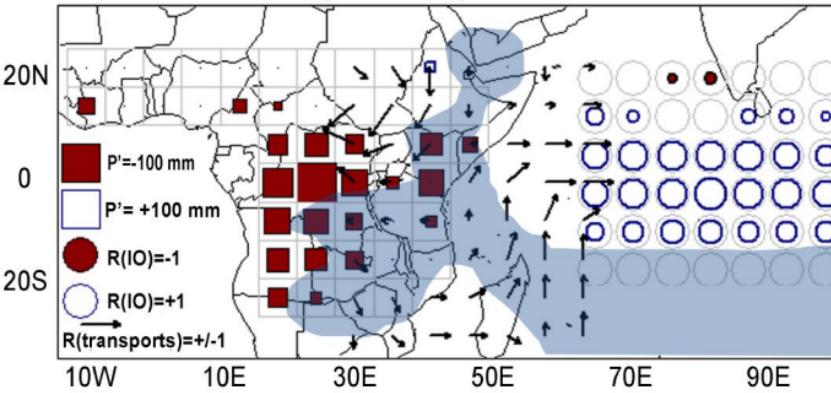
A. Indian Ocean sea surface temperatures and global surface temperatures tightly coupled and rising steeply since 1980



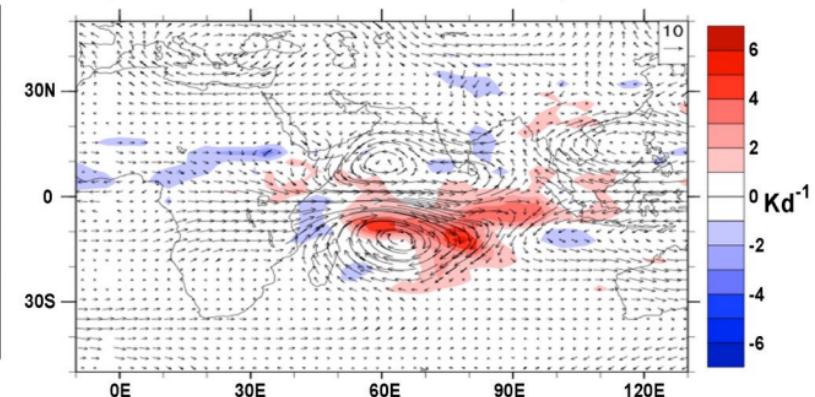
B. Regressions show negative relationships between growing season rainfall and Indian Ocean precipitation



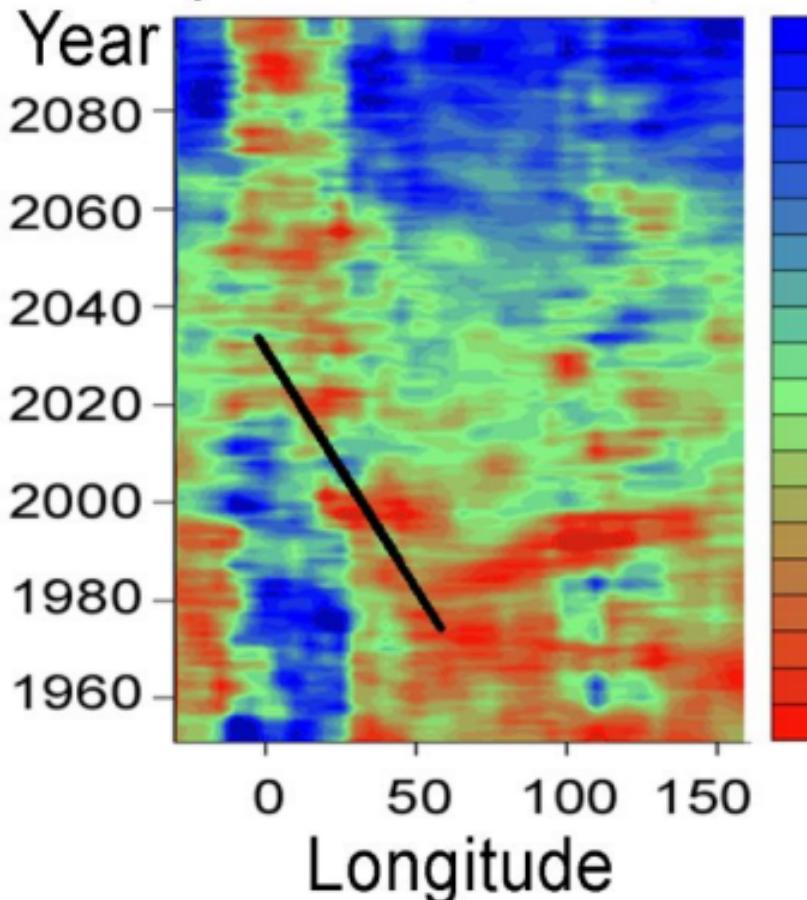
C. Canonical correlations link increasing oceanic precipitation with moisture transport disruptions



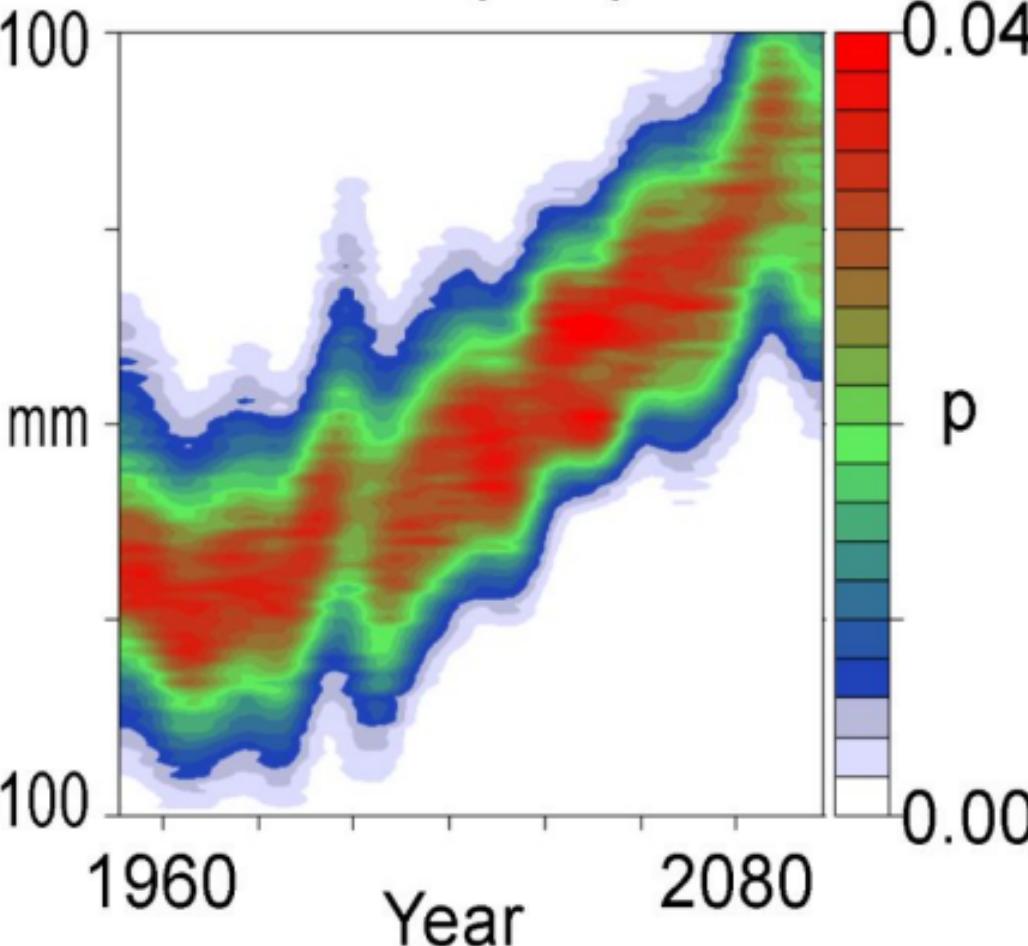
D. CAM diabatic forcing experiment results produce eastward moisture transports



A. Tropical rainfall z-scores



B. Indian Ocean precipitation PDFs



C. Individual Model Trends

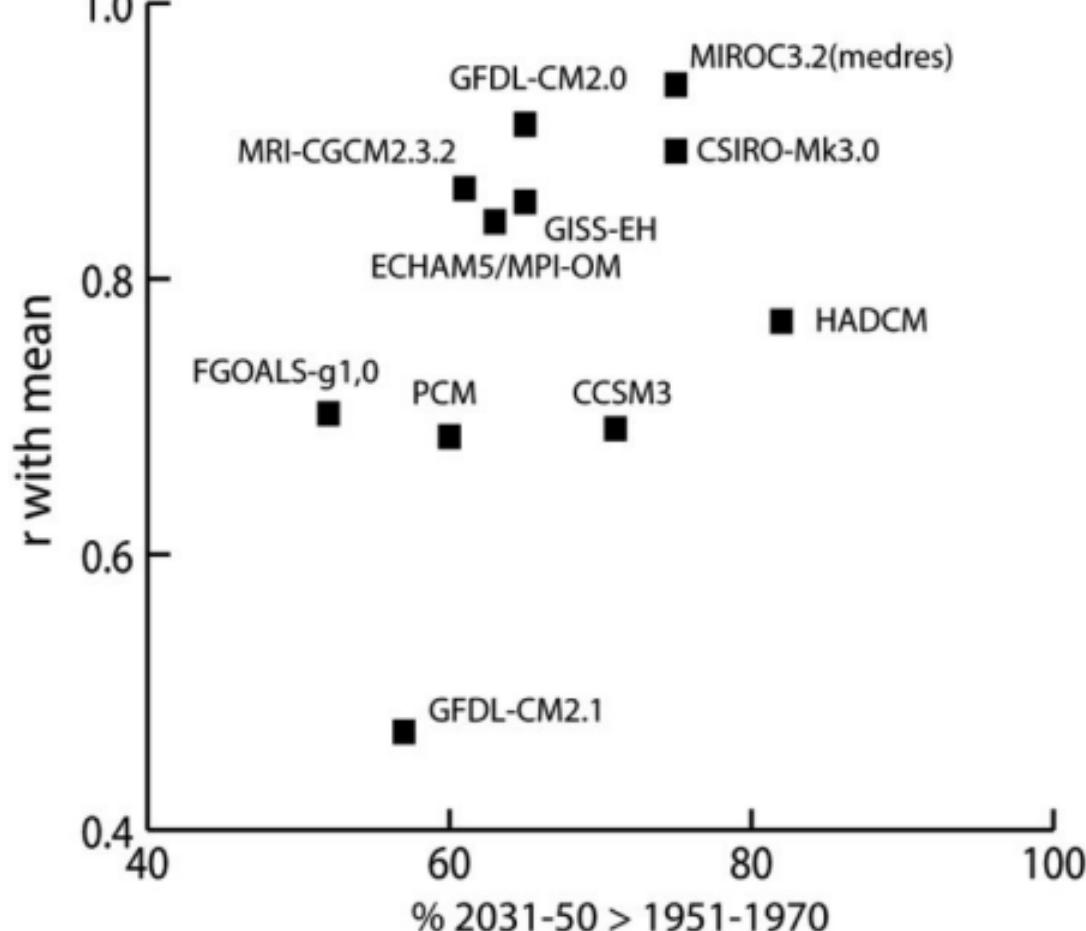


Table 1. Correlations between climate, rainfall and food security indices

	DJF R (p-value)	MAM R (p-value)	JJA R (p-value)
CC_P & CC_U			
Inter-annual (1950-05)	0.45 (5.6x10 ⁻⁴)	0.81 (6.5x10 ⁻¹⁴)	0.64 (1.4x10 ⁻⁸)
7-year timescale (1950-05)	0.90 (2.3x10 ⁻³)	0.97 (6.6x10 ⁻⁵)	0.90 (2.3x10 ⁻³)
CC_U & growing season rainfall	Southern Africa	Eastern Africa	Sahel countries
Inter-annual (1950-05)	-0.39 (3.2x10 ⁻³)	-0.50 (1.0x10 ⁻⁴)	-0.57 (2.7x10 ⁻⁵)
7-year timescale (1950-05)	-0.82 (4.6x10 ⁻²)	-0.92 (9.3x10 ⁻³)	-0.92 (9.3x10 ⁻³)
CC_P & CMIP3 oceanic rainfall			
7-year timescale (1966-05)	0.97 (1.2x10 ⁻³)	0.91 (1.6x10 ⁻²)	0.86 (2.8x10 ⁻²)
CC_P & Global Temperature			
7-year timescale (1966-05)	0.88 (2.0x10 ⁻²)	0.85 (3.2x10 ⁻²)	0.84 (3.6x10 ⁻²)